

1. Solution:

We want the value of overall chance of the failure, $1 - \prod_{i=1}^n (1 - (1 - r_i)^{b_i})$, to be as small as possible. Which

is equivalent to make $\prod_{i=1}^n (1 - (1 - r_i)^{b_i})$ as large as possible. So, this question is to find the optimal combination of the number of backup subsystems for every subsystem, and make their cost is less or equal to

B . Let $dp[0..B]$ is an array, $dp[i]$ denote that the maximal value of $\prod_{i=1}^n (1 - (1 - r_i)^{b_i})$ when our budget is i .

Then we can get the following recurrence formula:

$$dp[i] = (dp[i - c_k] \frac{1 - (1 - r_k)^{b_k + 1}}{1 - (1 - r_k)^{b_k}}) \quad \text{When } i - c_k \geq 0$$

In order to compute this value, we need to store the selection method of every budget i . That is, for every $dp[i]$, we will have a table $T_i[1..n]$ to store the number of every subsystem. That's where we can get our b_k on the

formula above ($b_k = T_i[k]$). Obviously, for $i = 0$, $T_0[1..n] = [0, 0, \dots, 0]$ and

$dp[0] = \prod_{i=1}^n (1 - (1 - r_i)^0) = 0$. So, we have initialized our base problem. We can then compute $dp[1]$,

$dp[2]$, ..., $dp[B]$. Of course, we must also keep track of the value of $T_i[1..n]$. After we compute the $dp[B]$.

The $T_B[1..n]$ will record the number of every subsystem that will get a maximal value of $\prod_{i=1}^n (1 - (1 - r_i)^{b_i})$

(the probability of success). Total running time is $O(Bn)$, because for every i , we need $O(n)$ time to compute the optimal combination and the optimal value $dp[i]$.

2. Solution:

```
Alg:uniquePaths(self, m: int, n: int) -> int:
    d = [[1...1]...[1...1]] # n * m
    for col->1 to m:
        for row->1 to n:
            d[col][row] = d[col - 1][row] + d[col][row - 1]
    return d[m - 1][n - 1]
```

3. Solution:

Suppose we know that a particular item of weight w is in the solution. Then we must solve the subproblem on $n - 1$ items with maximum weight $W - w$. Thus, to take a bottom-up approach we must solve the 0-1 knapsack problem for all items and possible weights smaller than W . We'll build an $n + 1$ by $W + 1$ table of values where

the rows are indexed by item and the columns are indexed by total weight. (The first row and column of the table will be a dummy row). For row i column j , we decide whether or not it would be advantageous to include item i in the knapsack by comparing the total value of of a knapsack including items 1 through $i - 1$ with max weight j , and the total

value of including items 1 through $i - 1$ with max weight $j - i.\text{weight}$ and also item i . To solve the problem, we simply examine the n, W entry of the table to determine the maximum value we can achieve. To read off the items we include, start with entry n, W . In general, proceed as follows: if entry i, j equals entry $i-1, j$, don't include item i , and examine entry $i-1, j$ next. If entry i, j doesn't

equal entry $i - 1, j$, include item i and examine entry $i - 1, j - i.\text{weight}$ next. See algorithm below for construction of table:

Algorithm 1 0-1 Knapsack(n, W)

```

1: Initialize an  $n + 1$  by  $W + 1$  table  $K$ 
2: for  $j = 1$  to  $W$  do
3:    $K[0, j] = 0$ 
4: end for
5: for  $i = 1$  to  $n$  do
6:    $K[i, 0] = 0$ 
7: end for
8: for  $i = 1$  to  $n$  do
9:   for  $j = 1$  to  $W$  do
10:    if  $j < i.\text{weight}$  then
11:       $K[i, j] = K[i - 1, j]$ 
12:    end if
13:     $K[i, j] = \max(K[i - 1, j], K[i - 1, j - i.\text{weight}] + i.\text{value})$ 
14:  end for
15: end for

```

4.

Many kinds of solutions. Here some examples:

Dp:

```

1  enum Index {
2      GOOD, BAD, UNKNOWN
3  }
4
5  public class Solution {
6      Index[] memo;
7
8      public boolean canJumpFromPosition(int position, int[] nums) {
9          if (memo[position] != Index.UNKNOWN) {
10             return memo[position] == Index.GOOD ? true : false;
11          }
12
13          int furthestJump = Math.min(position + nums[position], nums.length - 1);
14          for (int nextPosition = position + 1; nextPosition <= furthestJump; nextPosition++) {
15              if (canJumpFromPosition(nextPosition, nums)) {
16                  memo[position] = Index.GOOD;
17                  return true;
18              }
19          }
20
21          memo[position] = Index.BAD;
22          return false;
23      }
24
25      public boolean canJump(int[] nums) {
26          memo = new Index[nums.length];
27          for (int i = 0; i < memo.length; i++) {
28              memo[i] = Index.UNKNOWN;
29          }
30          memo[memo.length - 1] = Index.GOOD;
31          return canJumpFromPosition(0, nums);
32      }
33  }

```

Dp-bottom-up

```

1  enum Index {
2      GOOD, BAD, UNKNOWN
3  }
4
5  public class Solution {
6      public boolean canJump(int[] nums) {
7          Index[] memo = new Index[nums.length];
8          for (int i = 0; i < memo.length; i++) {
9              memo[i] = Index.UNKNOWN;
10          }
11          memo[memo.length - 1] = Index.GOOD;
12
13          for (int i = nums.length - 2; i >= 0; i--) {
14              int furthestJump = Math.min(i + nums[i], nums.length - 1);
15              for (int j = i + 1; j <= furthestJump; j++) {
16                  if (memo[j] == Index.GOOD) {
17                      memo[i] = Index.GOOD;
18                      break;
19                  }
20              }
21          }
22
23          return memo[0] == Index.GOOD;
24      }
25  }

```

Greedy:

```
1 public class Solution {  
2     public boolean canJump(int[] nums) {  
3         int lastPos = nums.length - 1;  
4         for (int i = nums.length - 1; i >= 0; i--) {  
5             if (i + nums[i] >= lastPos) {  
6                 lastPos = i;  
7             }  
8         }  
9         return lastPos == 0;  
10    }  
11 }
```